



Adaptive Density Estimation of Particle Data

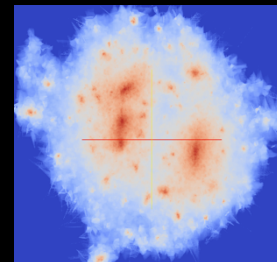
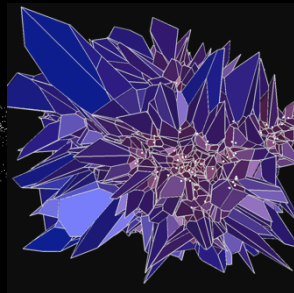
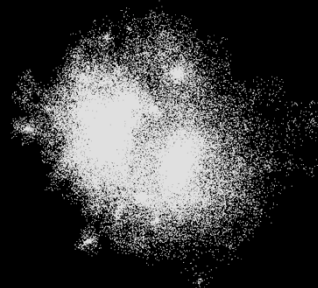
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Halo particles,
Voronoi
tessellation, and
2D density
estimation



CoDA Invited Talk
3/5/14

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Mathematics and Computer Science Division

Executive Summary

We describe work in progress for sampling a regular density field from a distribution of particle positions using a Voronoi tessellation as an intermediate data model.

Key Ideas

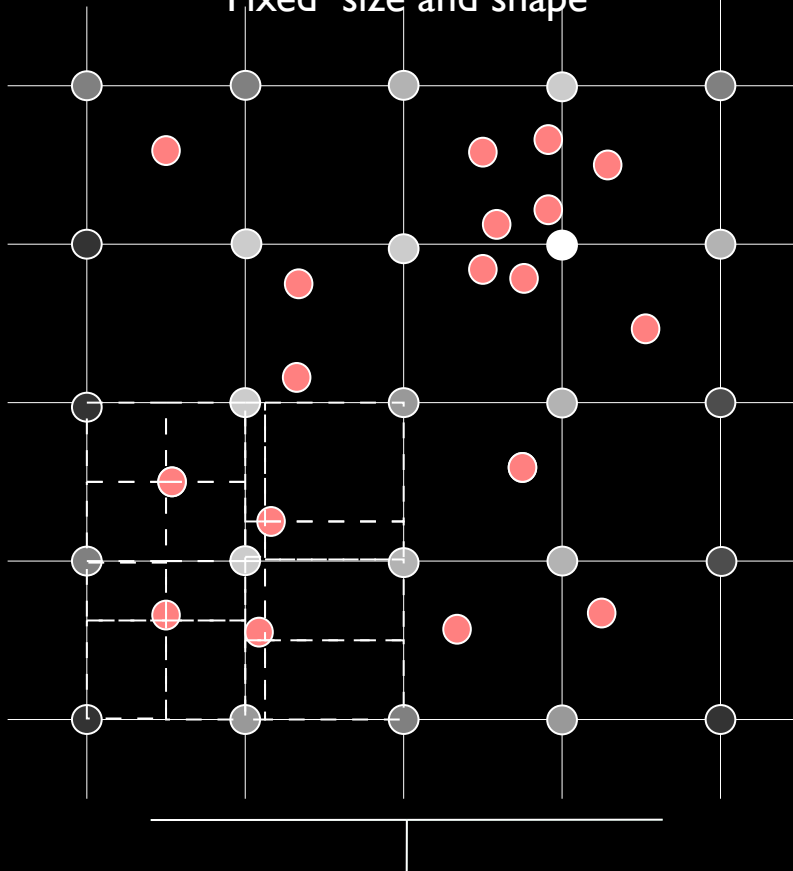
- Convert discrete particle data into continuous function that can be interpolated, differentiated, interpolated, represented as a regular grid (field)
- Automatically adaptive window size and shape
- Comparison with CIC and SPH using synthetic and actual data
- Voronoi tessellation and density estimation computed in parallel on distributed-memory HPC machines
- Application to gravitational lensing

Preliminaries

Estimation Kernels

CIC

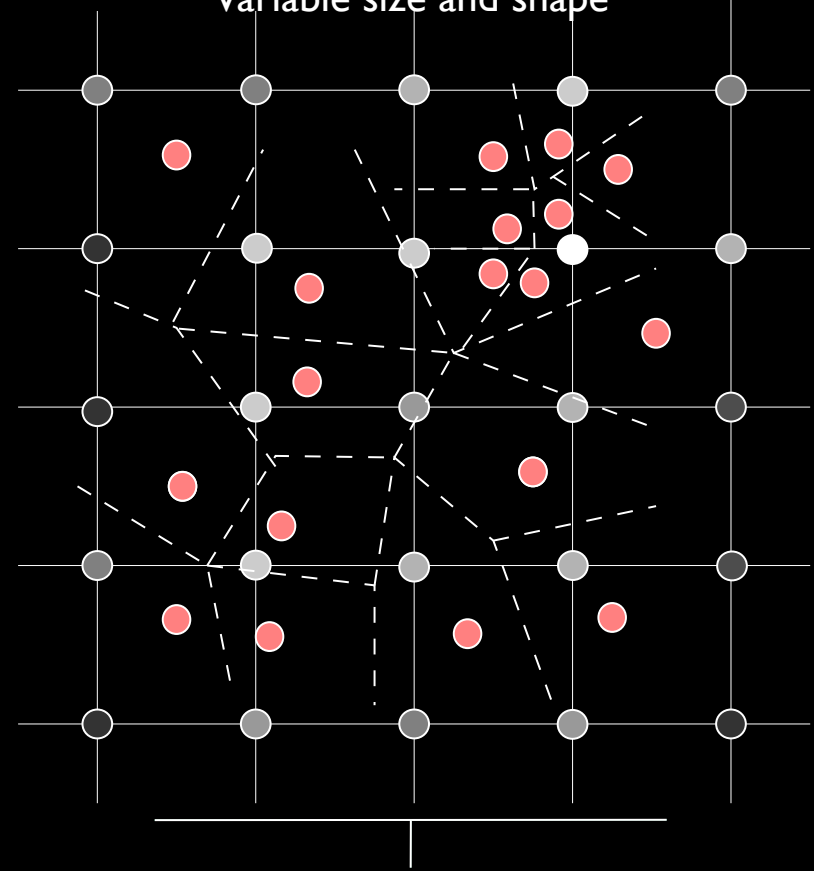
Fixed size and shape



In cloud-in-cell (CIC) methods, particles are distributed to a fixed number of grid points.

TESS

Variable size and shape

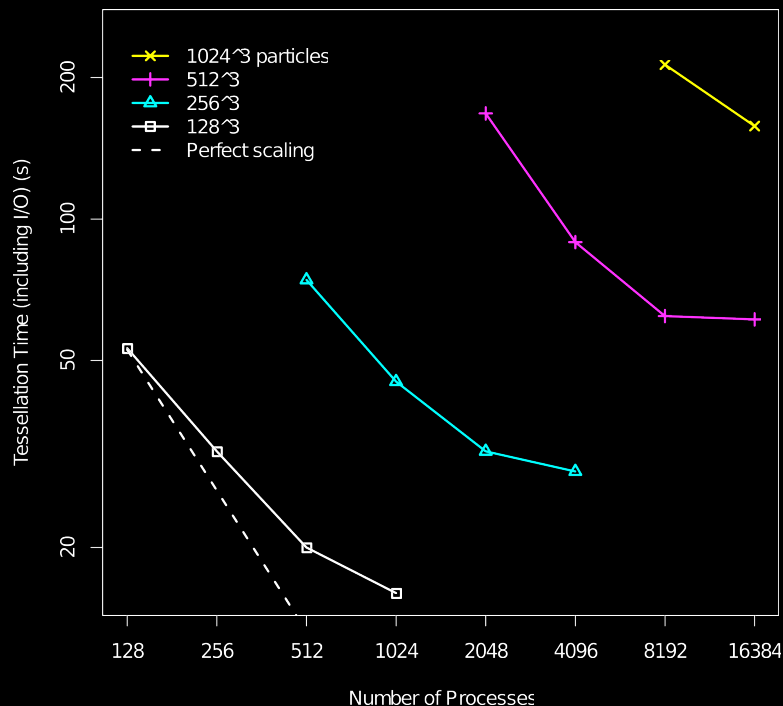


In tessellation (TESS) methods, particles are distributed to a variable number of grid points according to the Voronoi or Delaunay tessellation that has variable size and shape cells.

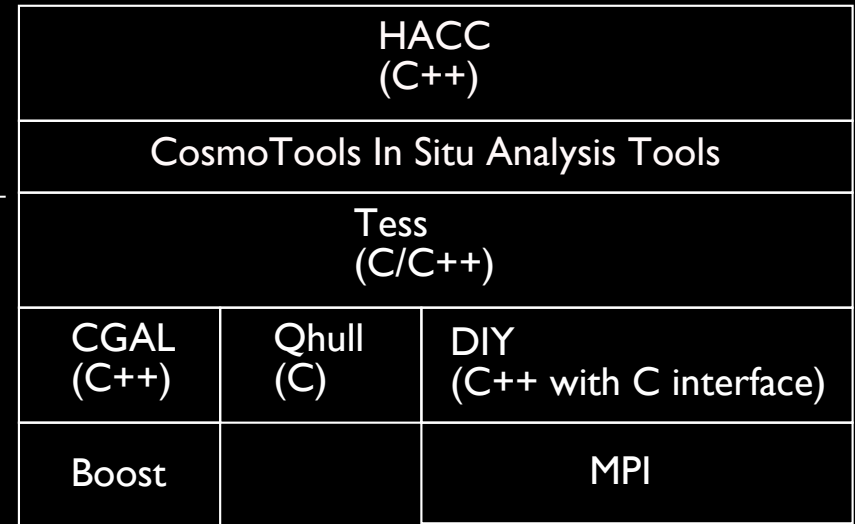
Tess Library

Tess is our parallel library for large-scale distributed-memory Voronoi and Delaunay tessellation.

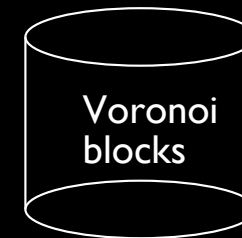
Strong Scaling



Dense, our density estimator, currently reads the tessellation from disk and estimates density onto a regular grid. Eventually dense will be converted to a library that can be coupled in memory to tess output, saving the tessellation storage.



Parallel NetCDF write



parallel read

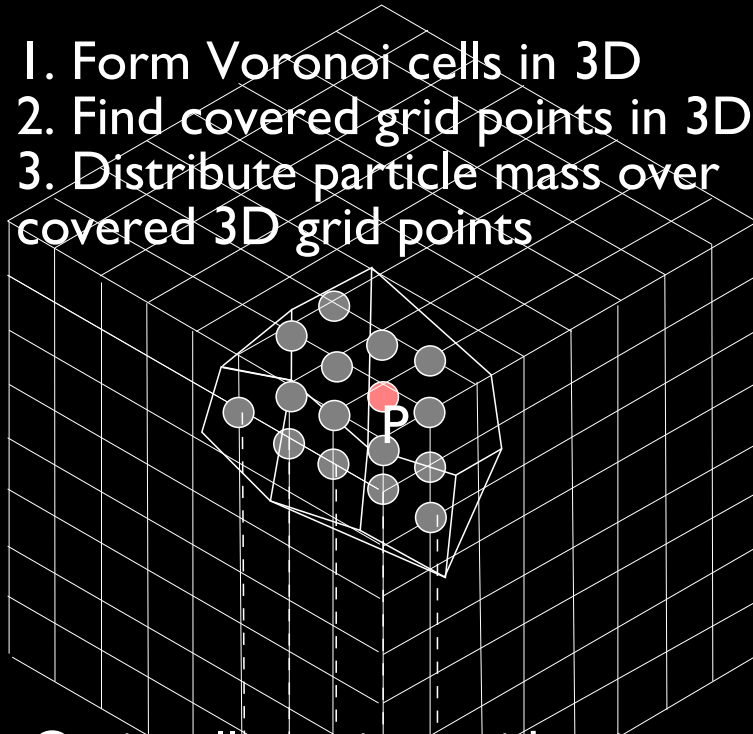
Postprocessing Cosmo Tools Plugin

ParaView

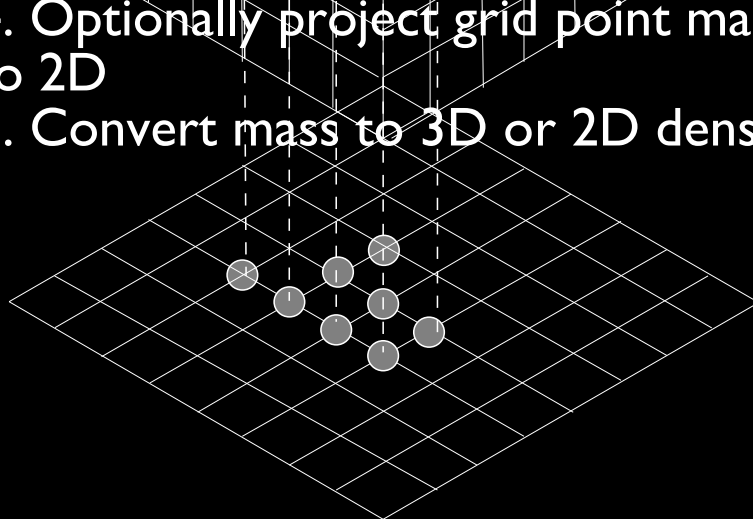
Method

Overall Algorithm

1. Form Voronoi cells in 3D
2. Find covered grid points in 3D
3. Distribute particle mass over covered 3D grid points



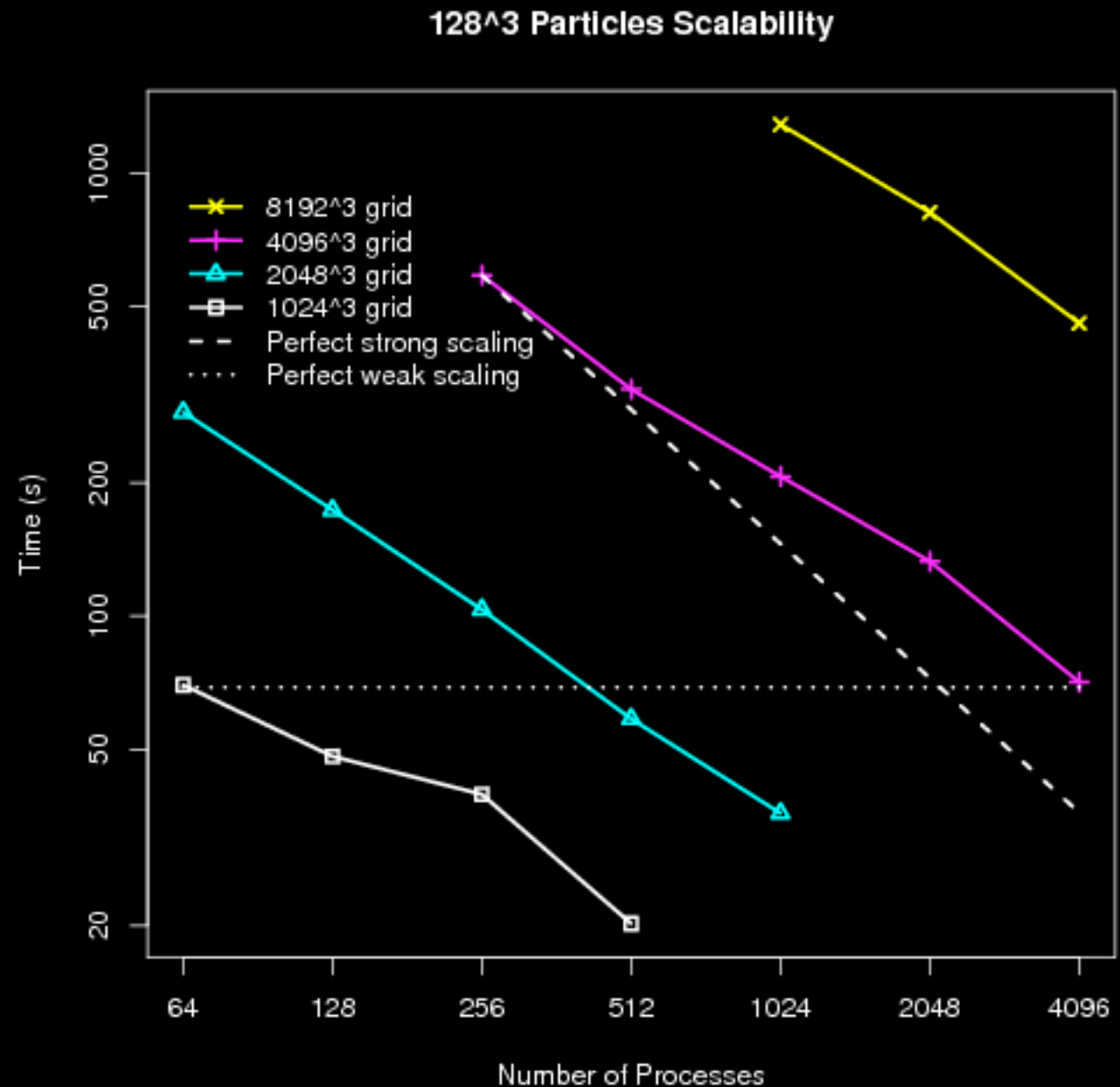
4. Optionally project grid point mass to 2D
5. Convert mass to 3D or 2D density



```
for (all Voronoi cells) {  
    compute grid points in cell bounding box  
  
    compute Voronoi cell interior grid points from  
        grid points in cell bounding box  
  
    for (all interior grid points) {  
  
        if (grid point is in bounds of local block)  
            add mass contribution to grid point  
  
        else  
            send mass contribution to neighboring block  
                containing grid point and add it there  
  
        if (no grid points in interior of Voronoi cell)  
            add mass contribution to single nearest  
                grid point  
  
        if (2D projection) {  
            accumulate mass at 2D pixel  
            divide by pixel area for 2D density  
        }  
  
        else  
            divide by voxel volume for 3D density  
  
    } // interior grid points  
}  
} // Voronoi cells
```

Dense Strong and Weak Scaling

- 128^3 synthetic particles
- End-to-end time (including reading tessellation and writing image)
- 3D->2D projection
- 51% strong scaling (End-to-end) for 4096^3 grid



Accuracy

Navarro-Frenk-White (NFW)

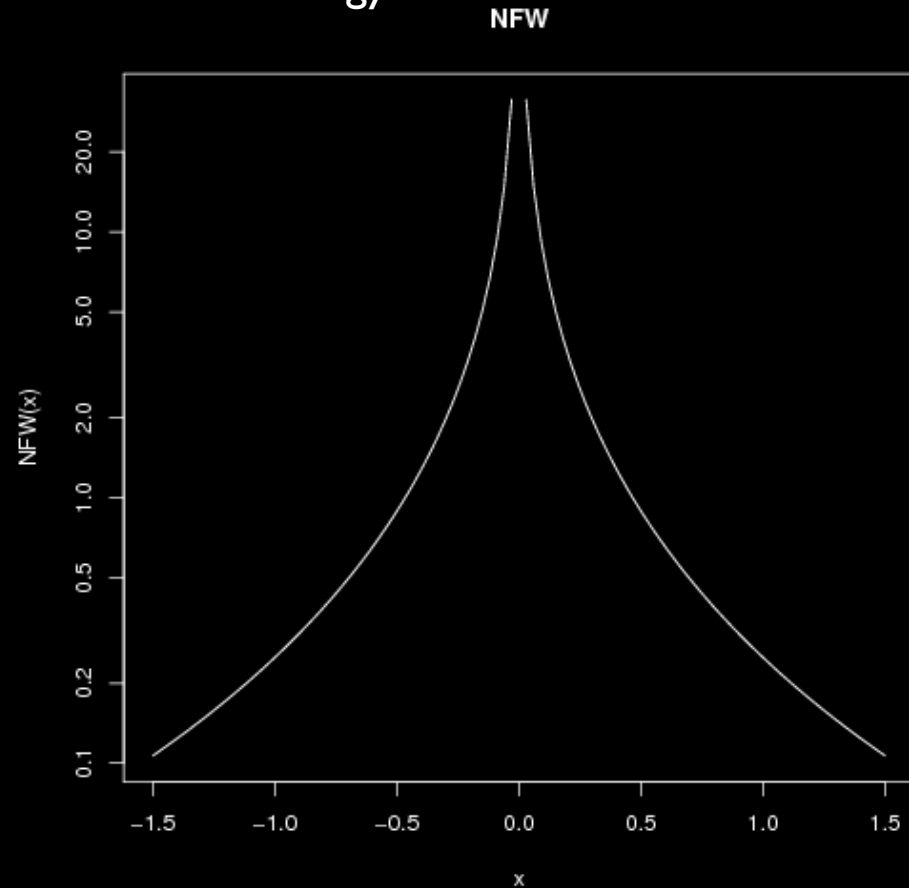
Our first synthetic dataset is derived from an analytical density function commonly used in cosmology.

k is a constant, 1 for us

$\rho(r)$ is Monte Carlo sampled to get test set of particles

Ground truth is 2D plot of $\rho(r)$

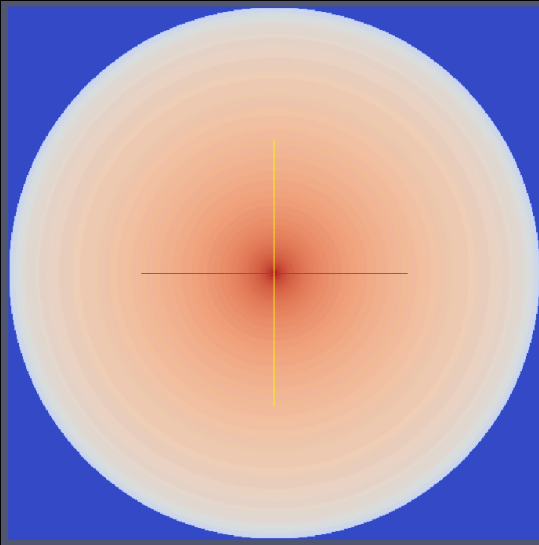
We limit r to $[-1.5, 1.5]$ and $\text{NFW}(r)$ to 10^6



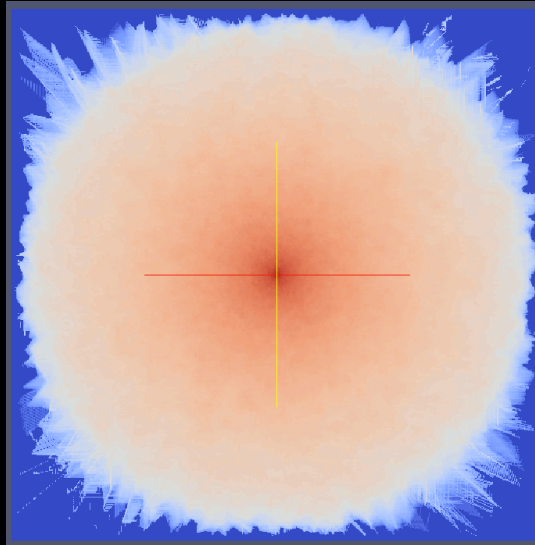
$$\rho(r) = \frac{k}{(r(r+1)^2)}$$

NFW 2D Density Fields

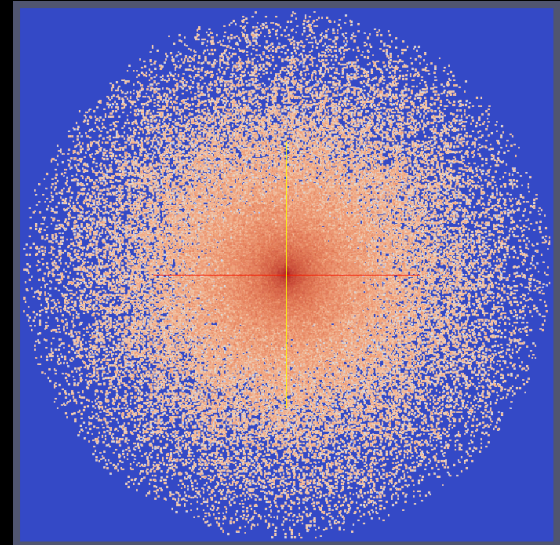
Analytical



TESS

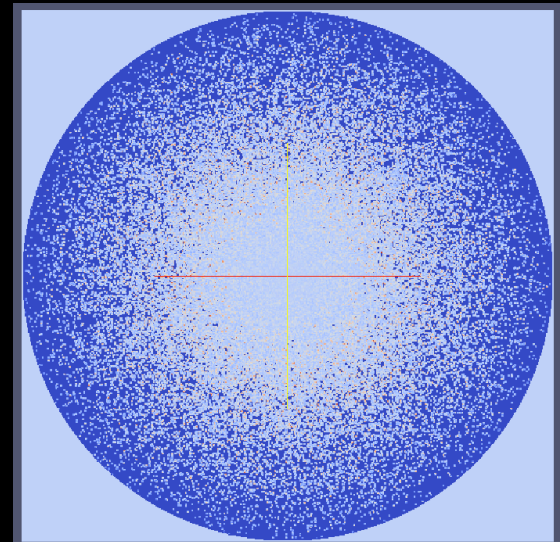
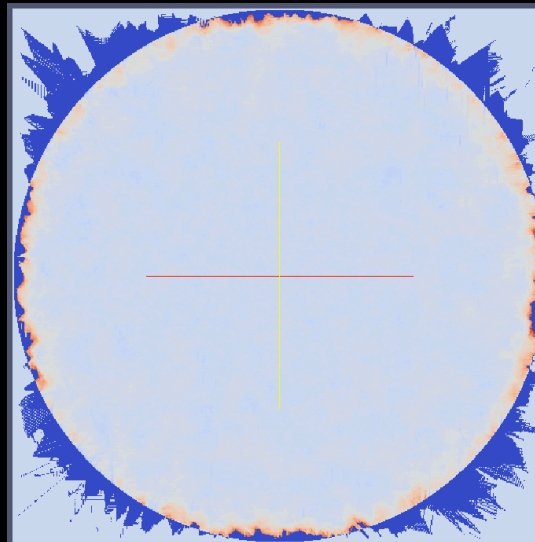


CIC



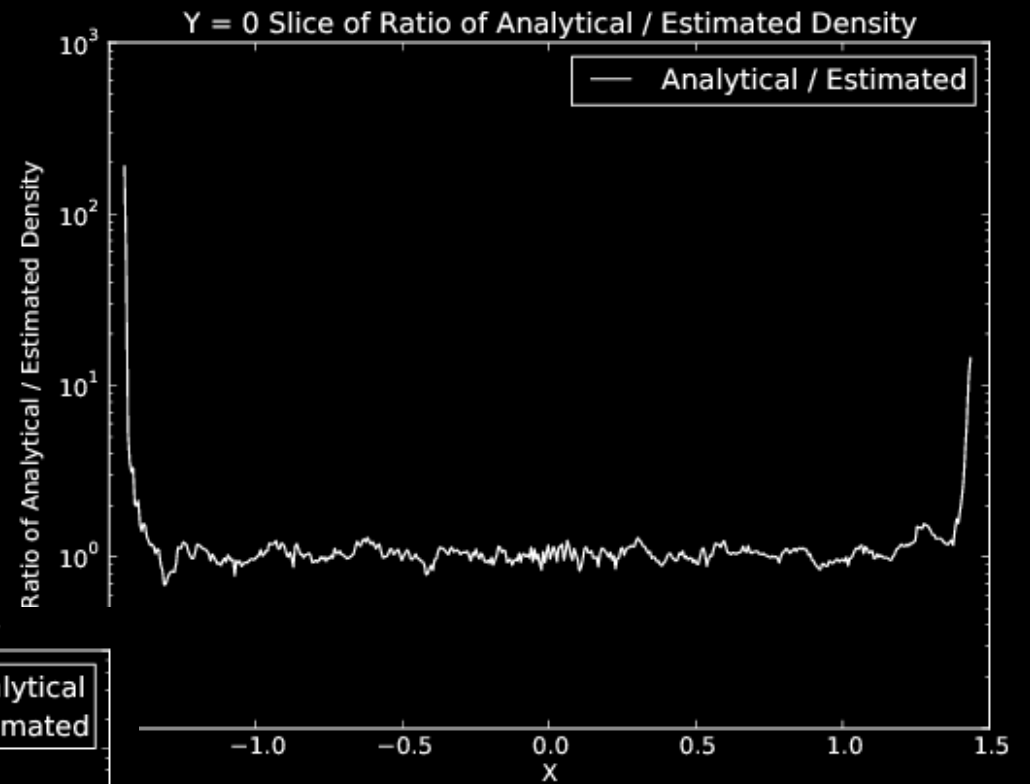
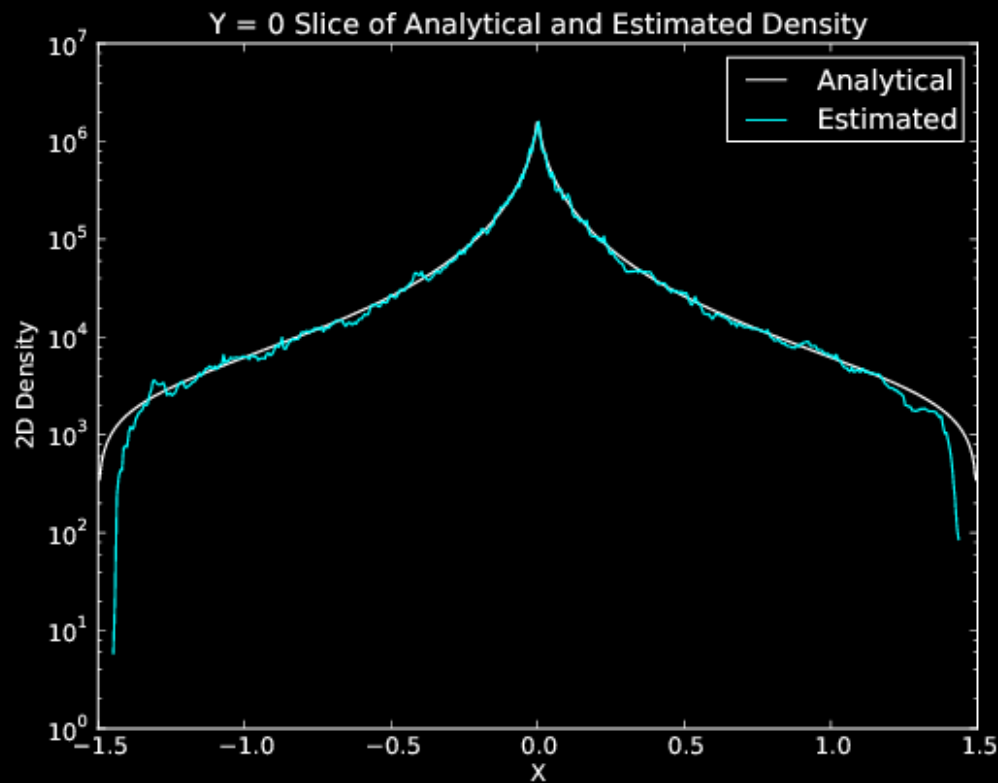
Top row:
 1024^3 3D density projected
to 1024^2 2D density field
and rendered in ParaView

Bottom row:
Ratio of analytical divided
by estimated density



TESS

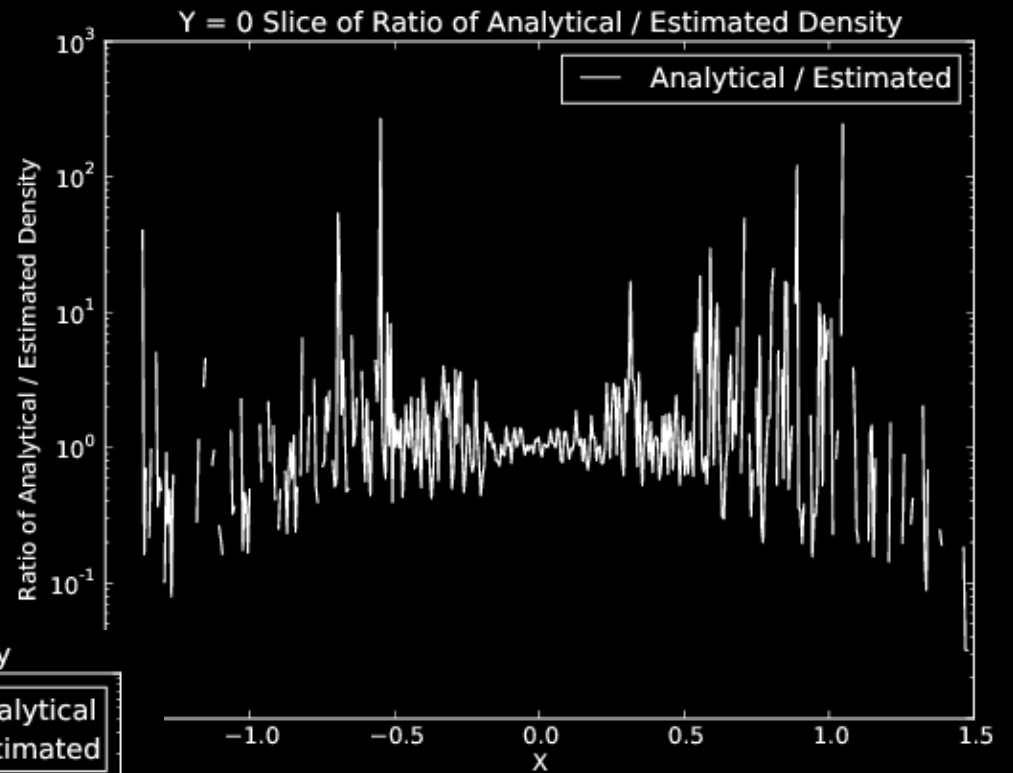
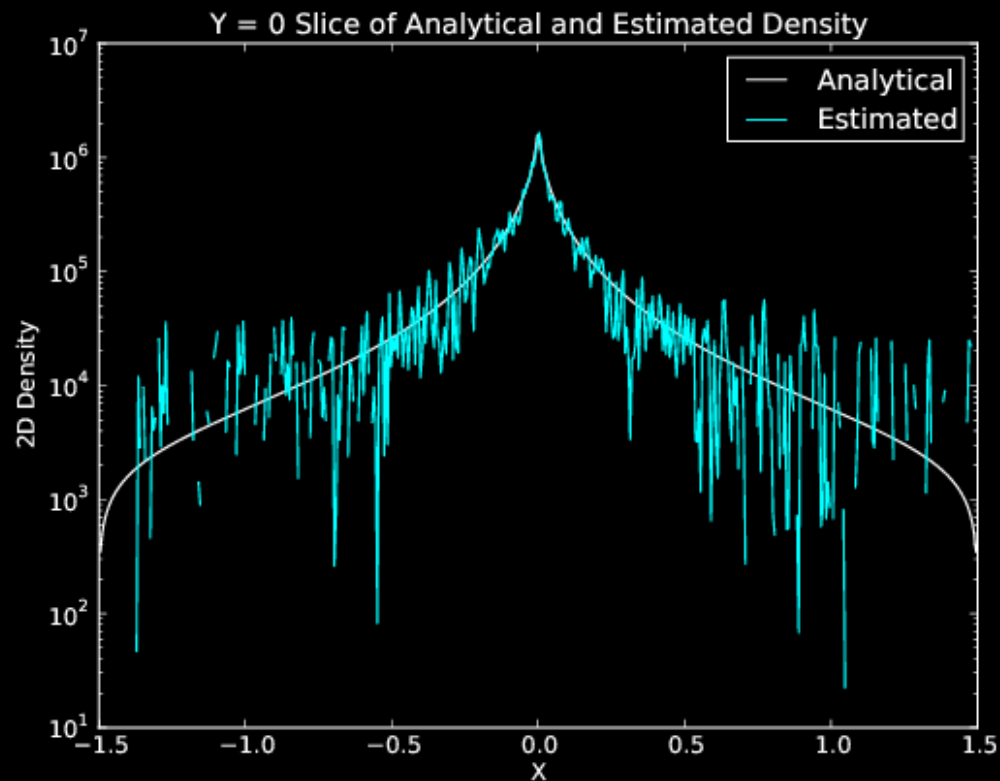
Comparison between analytical 2D density and estimated density at $y = 0$ cross section



Ratio between analytical 2D density divided by estimated density at $y = 0$ cross section

CIC

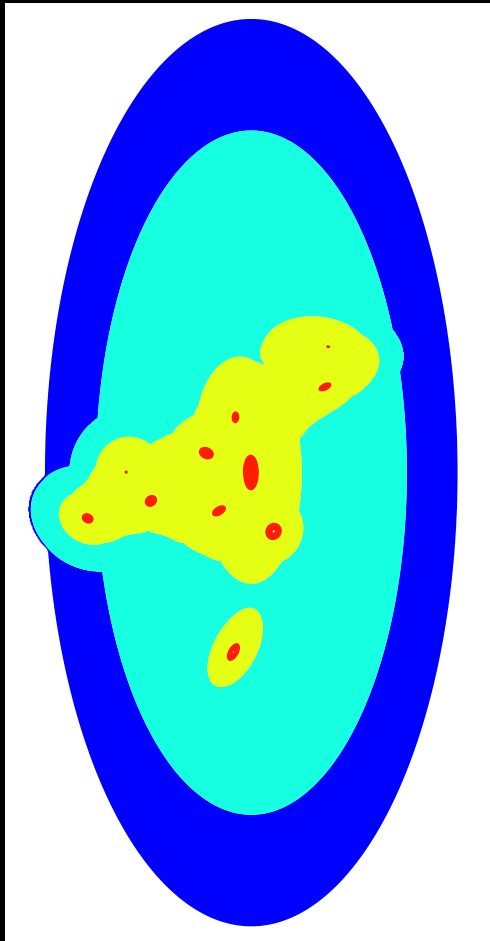
Comparison between analytical 2D density and estimated density at $y = 0$ cross section



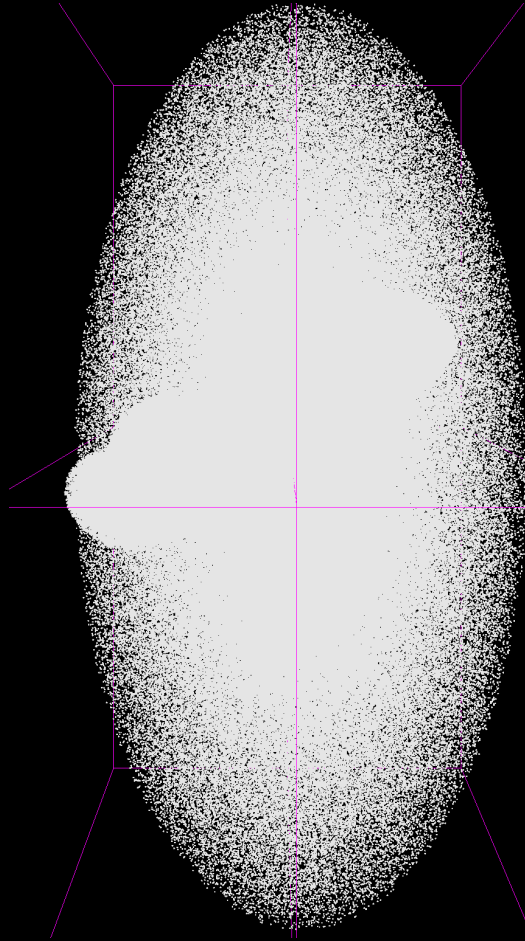
Ratio between analytical 2D density divided by estimated density at $y = 0$ cross section

Complex NFW (CNFW)

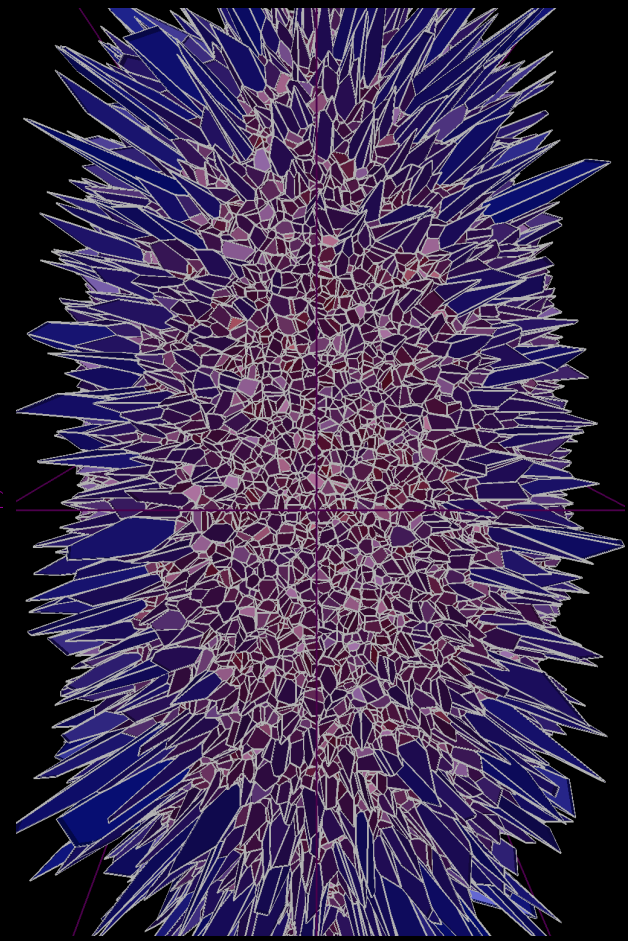
Our second synthetic dataset is a combination of several NFWs of varying cutoff densities and asymmetric scaling factors.



Analytical cutoff density
contours



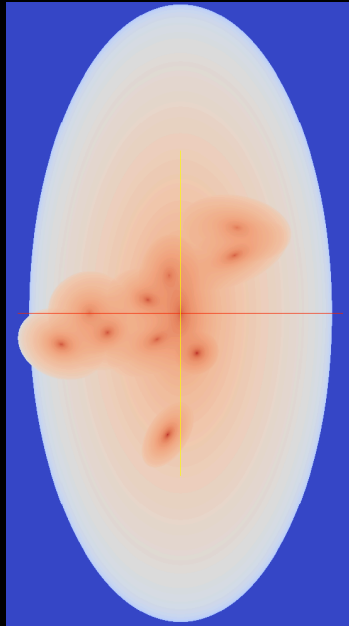
2e5 sampled particles



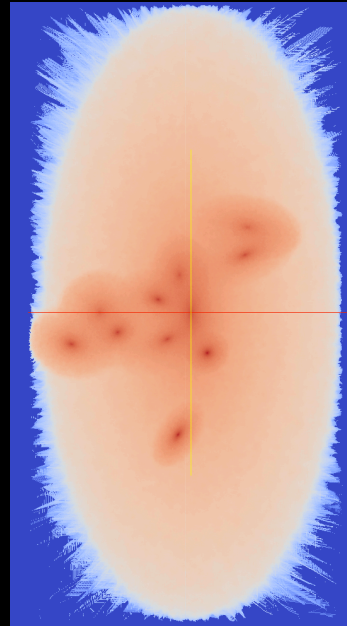
Voronoi tessellation

CNFW 2D Density Fields

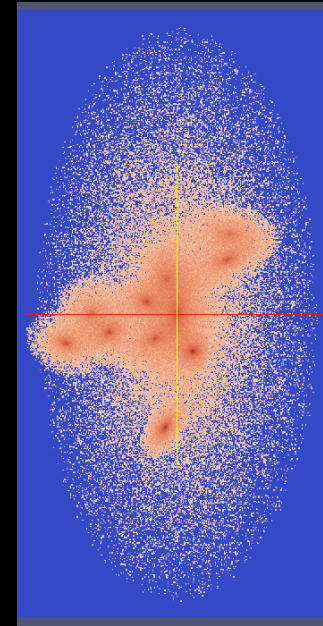
Analytical



TESS

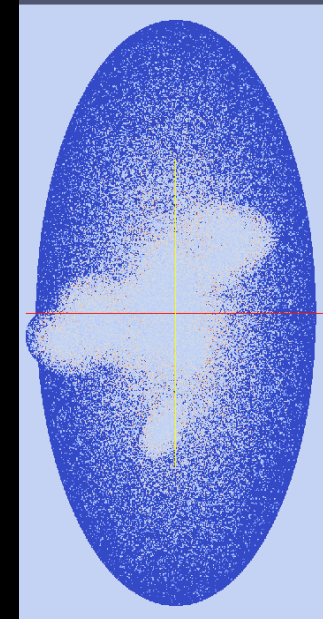
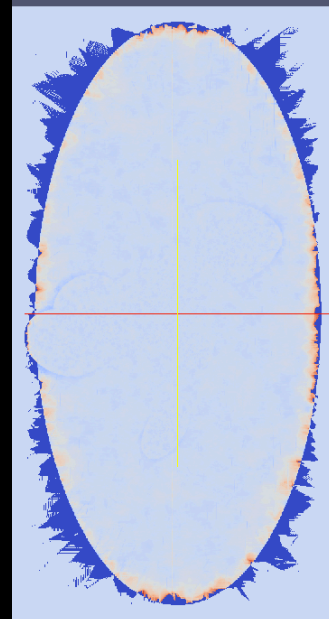


CIC



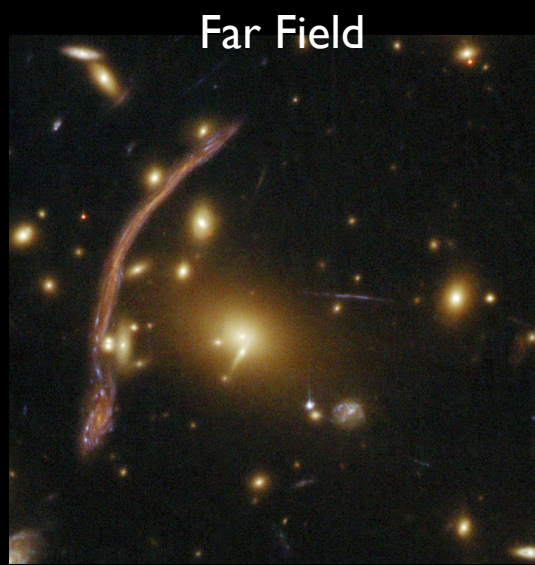
Top row:
 1024^3 3D density projected
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Ratio of analytical divided
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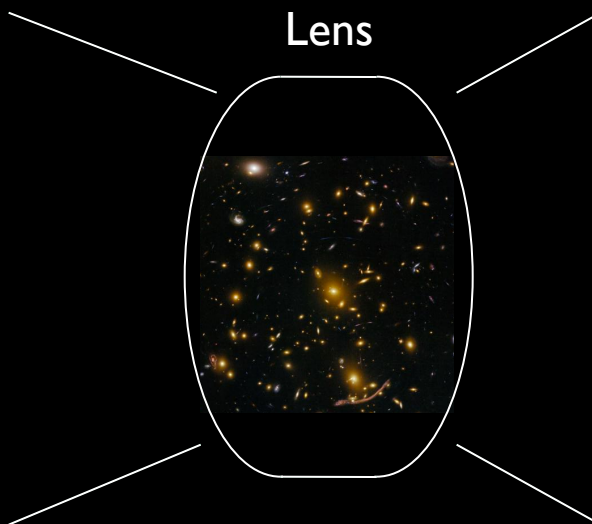


Application: Gravitational Lensing

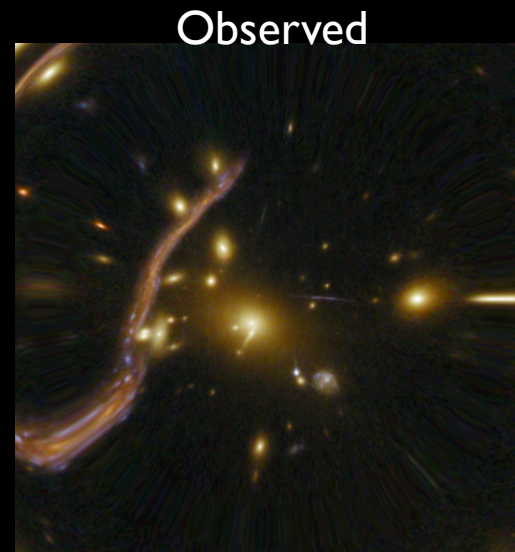
Lensing for Validating Simulations with Sky Surveys



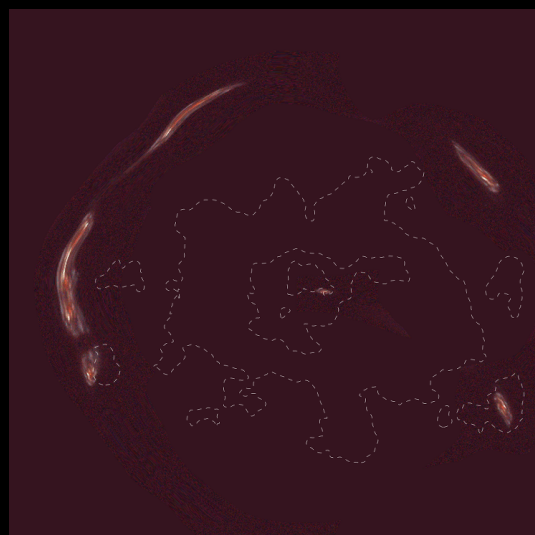
Actual far field



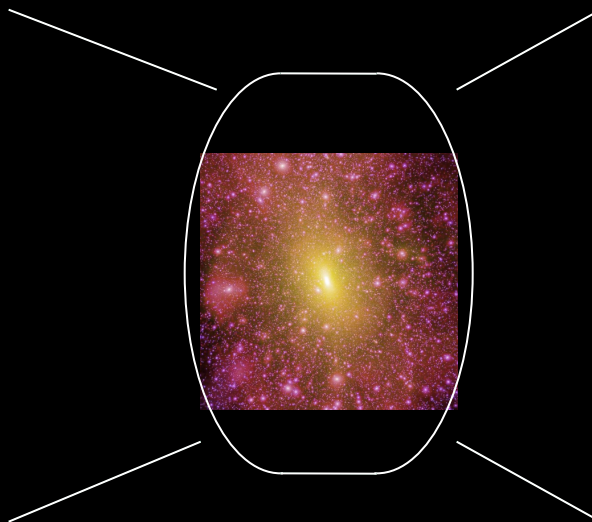
Actual near field density



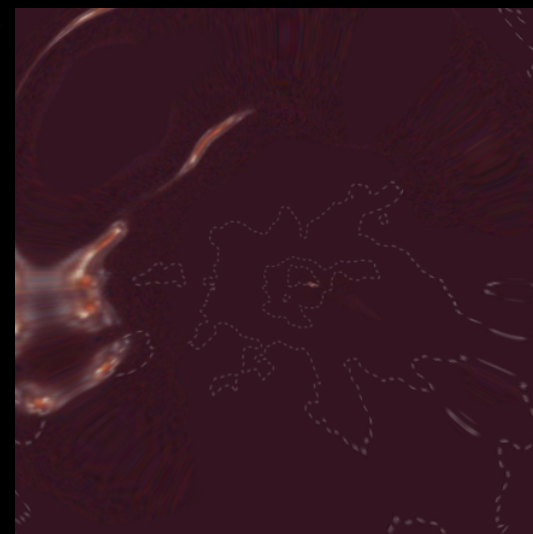
Distorted observations



Simulated far field

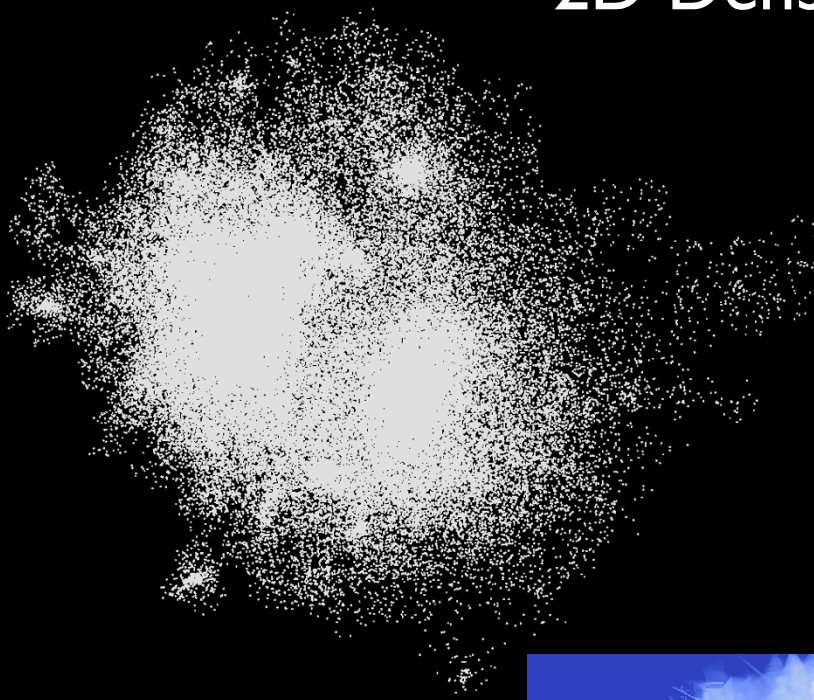


Simulated near field density

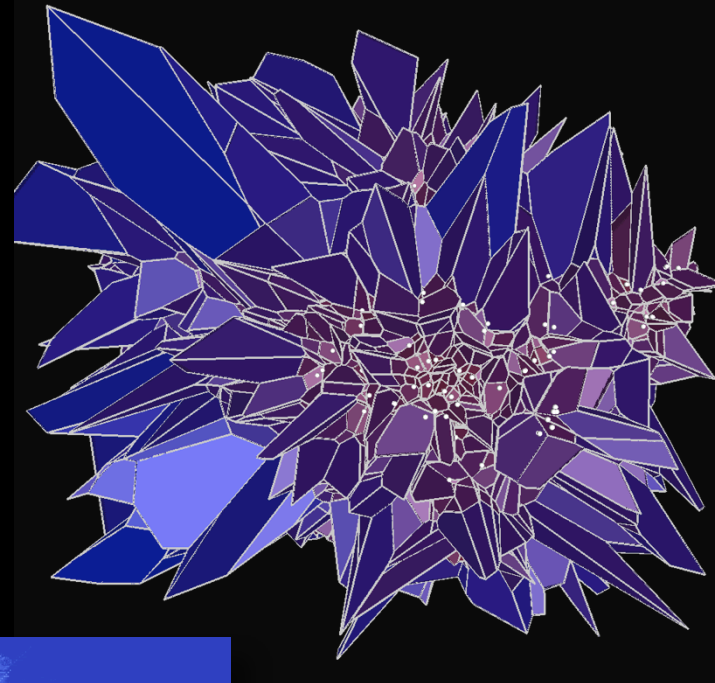


Simulated distortion

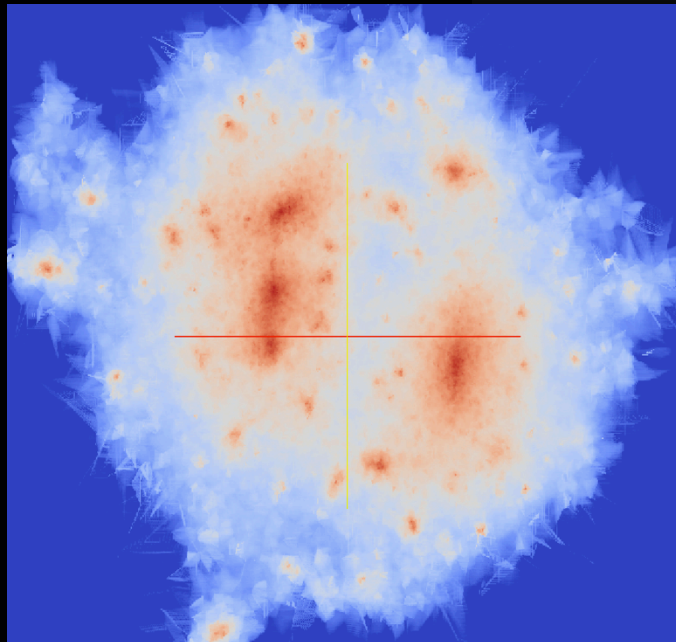
2D Density of Halo



Particle data from
HACC N-body
cosmology code from
halo ID 7445077095



Voronoi tessellation
of halo particles
colored by cell
volume



Final output
2D density
field for
lensing

Summary

I described sampling a regular density field from a distribution of particle positions using a Voronoi tessellation as an intermediate data model.

Key Ideas

- Automatically adaptive window size and shape
- Comparison with CIC and SPH using synthetic and actual data
- Voronoi tessellation and density estimation computed in parallel on distributed-memory HPC machines
- Application to gravitational lensing

Ongoing and Future Work

- Linear Barycentric interpolation inside Voronoi cells through Delaunay tessellation
- Shared memory threading inside MPI tasks
- Other applications such as 3D volume rendering

“The purpose of computing is insight, not numbers.”

–Richard Hamming, 1962

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People

Juliana Kwan, Hal Finkel, Adrian Pope, Nick
Frontiere, George Zagaris

Software

<https://repo.anl-external.org/repos/tess/trunk>

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